

# 3806ICT - Assignment 1

Nick van der Merwe - s5151332 - nick.vandermmerwe@griffithuni.edu.au

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## 1 Introduction

In this piece a controller is defined for a robot using a proportional integral derivative controller. This is split into two tasks, defining the PID algorithm and the Kalman filter. However, the architecture of the solution will contain four parts: the sonar wrapper, the PID service, the kalman filter service and the controller. Essentially, the sonar wrapper will turn the sonar topic into a service so that the controller does not subscribe to it. The PID and the Kalman filter will be purely functions without any storage so that they can be reused in the future. Lastly, the controller node will manage all of the data.

This report is organised based on the content in the nodes. Each section will introduce the necessary theory and explain how that is converted into code. With that in mind, Section 2 contains the sonar wrapper, Section 3 covers the PID equation (not defining the constants), Section 4 runs over the kalman filter (not finding the variance), and Section 5 defines the controller. The controller contains how all the previous nodes are tied together, including how the constants in PID are defined and variance is found for the Kalman filter.

## 2 Sonar Wrapper Node

This is quite straight forwards as it only requires us to save the last published value in a variable. As a design decision to avoid global variables it was made as a class. The code is visible in Figure 1a and 1b.

```
1 ---
2 assignment1_setup/Sonars readings
```

(a)

```
1 //
2 // sonarwrapper.cpp
3 // Copyright (C) 2021 Nick van der Merwe - nick.vandermmerwe@griffithuni.edu.au
4 //
5 // The purpose of this is to wrap the sonars topic and return
6 // what it says
7 //
8 #include "ros/ros.h"
9 #include "assignment1_setup/Sonars.h"
10 #include "assignment1_getSonarReadings.h"
11 #include <iomanip>
12
13 class SonarReader{
14 public:
15     SonarReader() = default;
16
17     void subCallback(const assignment1_setup::Sonars &msg){
18         lastReading = msg;
19     }
20
21     bool getSonarReadings(
22         assignment1_getSonarReadings::Request &req,
23         assignment1_getSonarReadings::Response &res){
24         res.readings = lastReading;
25         ROS_INFO("sending %d", (unsigned int)res.readings.distance);
26         return true;
27     }
28
29     private:
30         assignment1_setup::Sonars lastReading;
31 };
32
33 int main(int argc, char **argv){
34     ros::init(argc, argv, "sonar_wrapper_node");
35     ros::NodeHandle nodeHandle;
36
37     SonarReader sr{};
38     ros::Subscriber subSonarReader =
39         nodeHandle.subscribe("sonars", 1000,
40                               &SonarReader::subCallback, &sr);
41
42     ros::ServiceServer getSonarReadings =
43         nodeHandle.advertiseService("sonar_wrapper",
44                                     &SonarReader::getSonarReadings, &sr);
45
46     ROS_INFO("Ready to manage requests");
47
48     ros::spin();
49     return EXIT_SUCCESS;
50 }
```

(b)

Figure 1: Sonar wrapper srv in a and code in b

### 3 Task 1: PID Node

Before the coding section, the mathematics behind what a proportional integral derivative controller does must be explained. In short, the goal of the controller is to minimise the error that is reported. As stated in the task, the formula to be used here is in Equation 1.

$$\begin{aligned} Y(t) &= K_p e(t) + K_d \frac{de(t)}{dt} + K_i \int_0^t e(t') dt' \\ &= P(t) + D(t) + I(t) \end{aligned} \quad (1)$$

Where  $e(t)$  is the input error: in our case the distance from the bowl.  $t$  is the time,  $K_p$  is proportional gain,  $K_d$  is the derivative gain and  $K_i$  is the integral gain.

To make elaborating how this equation functions simpler, the first thing to do is to consider each of the terms separately as they each represent a part of **Proportional Integral Derivative**. To summarise, the proportional section represents the current error, the integral is the error in the past and the derivative is a prediction of the error into the future [M., 2000]. The logic behind this is that for the P term its just the current error, the integral is the area under the curve in the past and the derivative is the change at that specific point. These are each adjusted to the environment by using the K terms on each term.

Now the next issue is that this representation uses continuous time whereas in computational mathematics everything must be discrete - in this case our sonar only reports every 10ms. The exact maths for converting this into discrete values was covered in the lab content so it will not be copied over here, but rather explained.

The first step is to define time as discrete by making  $T = 10ms$  and  $t_i = T + t_i - 1$ . Technically this would be all that's necessary, but the exact way to calculate the derivative and integral is not known. For these the conceptual meanings can be used: a derivative is the change in the output of a function in the smallest unit of time and an integral is the area under the curve from one time to another. Another fact is that technically a derivative forwards (current minus future) is the same as backwards (current minus last value) in continuous time due to it being infinitely small.

To take this and define them discretely with logic is simple. A derivative is its current value subtracted by its last value, all divided by T (since that is the smallest unit of time we have). Next the integral is simply the sum of all the previous error values times T. This can conceptually be seen as how the area of a square is found at a low level without multiplication: split it into strips of size 1 (in our case T) and add these areas together. Including the K terms, our equations are in Equations 2, 3 and 4.

$$P(t) = K_p e(t) \quad (2)$$

$$D(t) = K_d \frac{e(t_i) - e(t_{i-1})}{T} \quad (3)$$

$$I(t_i) = K_i \sum_{i=1}^i T e(t_i) \quad (4)$$

Programmatically, it would be inefficient to do the integral by summing up everything in this manner every time it is calculated, so realistically the value is simply returned on its own in the function so that it can just be added to once in the next run. This is represented as Equation 5.

$$\begin{aligned} I(t_i) &= K_i f(t) \\ f(t) &= e(t_i)T + f(t_{i-1}) \end{aligned} \quad (5)$$

The PID srv file and C++ code is visible in figure 2.



```

1  float64 error
2  float64 lastError
3  float64 totalFValue
4  float64 K_p
5  float64 K_i
6  float64 K_d
7  float64 T
8  ---
9  float64 y
10 float64 lastError
11 float64 totalFValue

```

```

1  /*
2   * Written by Nick van der Merwe - s5151332
3   */
4  #include "ros/ros.h"
5  #include "assignment1_setup/Sonars.h"
6  #include "assignment1/pid_algorithm.h"
7  #include "geometry_msgs/Twist.h"
8  #include <cmath>
9  #include <algorithm>
10
11 /*
12  * We want to calculate the PID. To increase readability we
13  * are going to split each of these formulas into a separate function:
14  *  $y(t_i) = P(t_i) + I(t_i) + D(t_i)$ 
15  *  $P(t_i) = K_p * e(t_i)$ 
16  *  $I(t_i) = e(t_i) + I(t_{i-1})$ 
17  *  $D(t_i) = K_d * (e(t_i) - e(t_{i-1})) / T$ 
18  *  $e(t) = \text{distance from object at } t$ 
19  *  $T = \text{latency}$ 
20  */
21
22 bool pidAlgorithm(
23     assignment1::pid_algorithm::Request & req,
24     assignment1::pid_algorithm::Response & res){
25     double P = req.K_p * req.error;
26     res.totalFValue = req.error * req.T + req.totalFValue;
27     double I = req.K_i * res.totalFValue;
28     // split D into top and bottom so its more readable
29     double && topD = req.error - req.lastError;
30     double D = req.K_d * (topD / req.T);
31
32     res.y = P + I + D;
33     // checking the documentation 22.0 is the maximum speed
34     res.y = std::min(res.y, 0.22);
35     res.y = std::max(res.y, 0.0);
36
37     ROS_INFO("lastError: %u totalFValue %lf",
38             (unsigned int) res.lastError, res.totalFValue);
39     ROS_INFO("K_p: %lf K_i %lf K_d %lf", req.K_p, req.K_i, req.K_d);
40     ROS_INFO("P: %lf, I: %lf, D: %lf", P, I, D);
41     ROS_INFO("Returning y as %lf", res.y);
42     return true;
43 }
44
45 int main(int argc, char **argv){
46     ros::init(argc, argv, "pid_algorithm_node");
47     ros::NodeHandle nodeHandle;
48
49     ros::ServiceServer getSonarReadings =
50         nodeHandle.advertiseService("pid_algorithm", pidAlgorithm);
51
52     ROS_INFO("Ready to manage requests");
53     ros::spin();
54
55     return EXIT_SUCCESS;
56 }

```

Figure 2: The PID srv and cpp files respectively

## 4 Task 2: Kalman Filter Node

The goal of a Kalman filter is to get rid of noise in sonar readings through maths, and in this case odometry readings. The equations to be used are defined in Equations 6.

$$\begin{aligned} K_i &= \frac{P_i^-}{P_i^- + R_i} \\ y_i &= y_i^- + K_i(z_i - y_i^-) \\ P_i &= (1 - K_i)P_i^- \end{aligned} \tag{6}$$

Where  $z_i$  is the sonar's reading,  $y_i^-$  is the estimation of how far the robot is from the bowl at step  $i$ ,  $P_i^-$  is the predicted variance at  $i$ , and  $R_i$  is the initial variance read. Since there are no continuous variables in this, there is no need for this to be changed such as the PID. However, each of these should still be properly explained.

The naive way to solve the issue of a normally distributed noisy sonar would simply be to measure the median every time. However, to do that the robot would have to sit there for several seconds, move forwards slightly then sit there and measure the median again. Instead, what the Kalman filter aims to do is take the measuring once, move, (in our case) consider how far it moved through the odometry, apply the Kalman filter to this and get an estimate of how far it is [Siegward et al., 2011]. The reason why the odometry alone would not be enough is visible in how the sonar functions. Consider the case where the bowl lies 13 degrees to the right of the way the robot is facing. If we only use the euclidean distance measured by the odometry system, it would not actually be the distance from the bowl. From all of this, we can see the need for why the Kalman filter is necessary.

For now the focus is on applying the Kalman filter step out of the steps detailed previously which is all in equation 6. In short, the goal of this step is to give readings varying weight depending on the certainty of them. This weight is given by the Kalman gain by taking the predicted variance divided by itself added to the overall variance. Then we can add the difference in the reading and the estimated position to its estimated position to find  $y_i$ . To calculate the next predicted variance is typically complicated, but was compressed into multiplying  $P_i^-$  by the complement of the Kalman gain  $(1 - K_i)$ . In practice this just makes later readings have less weight.

In an attempt to understand this particular equation, an interesting proof by induction was found that makes calculating the variance irrelevant. Essentially it's possible to remove  $P_i$ ,  $P_i^-$  and  $R_i$  from the equations. First, let's change the notation so that the pattern is more obvious.

$$\begin{aligned} K_i &= \frac{P_i}{P_i + R} \\ P_{i+1} &= (1 - K_i)P_i \\ P_0 &= R \end{aligned}$$

First we define our base case of  $K_0$

$$\begin{aligned} K_0 &= \frac{P_0}{P_0 + R} \\ P_0 &= R \\ K_0 &= \frac{P_0}{2P_0} \\ K_0 &= \frac{1}{2} \end{aligned}$$

Now in induction we just have to prove the  $i + 1$  case

$$\begin{aligned}
 P_{i+1} &= R(1 - K_i) \\
 K_{i+1} &= \frac{P_{i+1}}{P_{i+1} + R} \\
 &= \frac{R(1 - K_i)}{R(1 - K_i) + R} \\
 &= \frac{R(1 - K_i)}{R - RK_i + R} \\
 &= \frac{R(1 - K_i)}{2R - RK_i} \\
 &= \frac{R(1 - K_i)}{R(2 - K_i)} \\
 &= \frac{1 - K_i}{2 - K_i}
 \end{aligned}$$

So essentially we found that

$$\begin{aligned}
 K_0 &= \frac{1}{2} \\
 &= \frac{1 - K_i}{2 - K_i}
 \end{aligned}$$

With this being said, it will just be ignored. It's an interesting proof that makes variance useless, but taking it out will probably remove a section of the assignment that the assignment writer intended the student to do. It's essentially just evidence of extensive research around understanding the formulas at work.

Moving onto the implementation, its a service that uses the exact formulas on the task sheet visible in figure 3.

```

1 float64 R_i
2 uint16 z_i
3 float64 y_i_estimate
4 float64 P_i_estimate
5 ---
6 float64 y_i
7 float64 P_i

```

(a)

```

1 //
2  * By Nick van der Merwe - s5151332
3  */
4 #include <limits>
5 #include <robo_msgs/ModelState.h>
6 #include "ros/ros.h"
7 #include "assignment1/kalman_filter.h"
8
9 [...]
10
11 bool kalmanFilter(
12     assignment1::kalman_filter::Request &req,
13     assignment1::kalman_filter::Response &res){
14     double K_i = req.P_i_estimate / (req.P_i_estimate + req.R_i);
15     res.y_i = req.y_i_estimate + K_i*(req.z_i - req.y_i_estimate);
16     ROS_INFO("P_i_estimate (before) %lf", req.P_i_estimate);
17     req.P_i_estimate = (1 - K_i) * req.P_i_estimate;
18     ROS_INFO("K_i %lf R_i %lf z_i %u y_i_estimate %lf P_i_estimate(after) %lf",
19         K_i, req.R_i, (unsigned int)req.z_i, req.y_i_estimate, req.P_i_estimate);
20
21     ROS_INFO("Returning y_i as %lf", res.y_i);
22     return true;
23 }
24
25 int main(int argc, char **argv){
26     ros::init(argc, argv, "kalman_filter_node");
27     ros::NodeHandle nodeHandle;
28
29     ros::ServiceServer findFilteredOutput =
30         nodeHandle.advertiseService("kalman_filter", kalmanFilter);
31
32     ROS_INFO("Ready to manage requests");
33     ros::spin();
34
35     return EXIT_SUCCESS;
36 }

```

(b)

Figure 3: Kalman filter implementation, srv file in 3a and code in 3b

## 5 Controller Node

### 5.1 PID constants

As picking the PID terms is not an essential step and only needs to be justified, we can use manual selection. To define our PID terms we must consider the goal of how we wish the robot to move. A reasonably logical approach would be to make the robot run at maximum speed until it comes close to the bowl and then slow down when it's closer so that it can accurately move. To align this with a real world goal, it's wanting to drive to a location as fast as possible yet be safe. In other words, when a car is parking it slows down so it can move more accurately.

One issue to consider is that typically the error in an equation is not just distance, which is especially impactful to the integral term. Usually when the error hits zero, the integral value is set as a constant otherwise it would continue to rise as the controller is running [Siegward et al., 2011]. However, this only happens when the robot actually reaches the bowl which would be the end of the problem. In the case that the robot is as far away as its sonar readers allow (around 60,000), it would likely mean that it will just run at maximum speed all the way to the bowl. This results in three options: 1) Set  $K_i$  to 0, rendering the entire term useless 2) Let the robot just run at maximum speed when it's far away 3) Decide on a point when the integral term should stop changing.

To be true to the goal, the best option here would be option 1. Option 2 is in direct conflict with the goal of slowing down when close as it renders it impossible in some situations, and picking an arbitrary point where the integral stops growing would only cause the robot to have a minimum constant speed after a certain point.

Another factor is the the D term also does the opposite of the goal of the robot. Using it we can either make the robot go faster when it's already going fast, or do the opposite. If this term were to dominate the terms it would mean that once the robot has hit a certain speed it will stay at that speed unless slowed by an outside force. Another option is that it will go in the opposite direction of where it is going, causing it to move back and forth (derivative is positive, move negative, derivative is negative, move positive... etc). As we do not want either of these, the derivative term can be set to 0 as well.

To tie it together, the derivative and integral are in direct conflict to the goal of getting there as fast as possible safely, and the only thing remaining is the proportional term. Ideally, the robot simply runs at maximum speed until a certain point which it starts slowing down to approach the bowl. To decide on this point the robot was placed reasonably close to the bowl and the distance was measured as in figure 4.

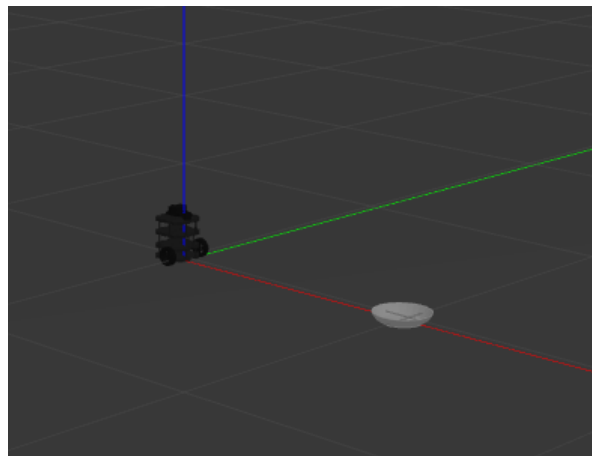


Figure 4: How far the robot looks when it's 100 units away

To work out our  $K_p$  from this we just have to find what the speed should be when it's at 100 units by doing  $y(t) = K_p e(t) \Rightarrow 0.22 = K_p 100 \Rightarrow K_p = 0.22/100$ . We can just set  $K_p$  to this.

## 5.2 Calculating Variance

To do this is not any different from usual:

$$\begin{aligned} \text{mean} &= \frac{1}{n} \sum_{i=0}^n x_i \\ \text{variance} &= \frac{\sum_{i=0}^n (x_i - \text{mean})^2}{n - 1} \end{aligned} \quad (7)$$

where  $x_i$  is a collected sample and  $n$  is the number of samples [Bhandari, 2020]. Note that the sample formula is used instead of the population formula. With regards to the number of samples collected, it was decided that 1000 will be used. Realistically, this is hard to justify without other variables involved but at 1000 points it is expected to give ten points for every percentile which should be reasonable.

One option is to hard-code in the variance because in the *sonars.cc* source code provided *std::normal\_distribution()* is used and the standard deviation is defined as 100. Meaning, the variance is hard-coded to  $100^2 = 10,000$ . However, hard-coding in the answer would destroy an intended part of the assignment and reduce reusability in the case that the sonar reader's distribution changes. Furthermore, this can be used as justification for why the variance is calculated every run: to improve reusability.

## 5.3 Structure

A core factor that makes implementing this difficult is that extracting functions from main in ROS is often prohibited due to a couple of methods only being permitted to be called inside main. So for easier understanding, figure 7 was made to grasp the implementation better.

```
// Created by Nick on 2/20/2021.
// https://www.lighthouse3d.com/roscpp/
// Nick_vandermerwe@griffith.com

const double UNITS_PER_RADIAN = M_PI / 180 * 36;

void defineInitialValues(geometry_msgs::Twist& twist, std::shared_ptr<SonarReadingSrv> sonarReadingSrv) {
    // We need to rotate
    if (sonarReadingSrv->response->readings->distance == UNITS_MAX) {
        sonarReadingSrv->response->readings->distance = UNITS_MAX;
    }
    // Change to the angle the robot is to just turn right
    double angle = 0;
    if (sonarReadingSrv->response->readings->distance == UNITS_MAX) {
        // If we need to increase distance it is we need to turn positive
        double angle = 0;
    } else {
        // If we need to decrease distance it is we need to turn negative
        double angle = 0;
    }
    twist->angular.z = UNITS_PER_RADIAN * angle;
}

void definePIDInitialValues(assignments::pid_algorithm& pidAlgorithm) {
    // PID control is used to keep the robot moving
    pidAlgorithm->request->K_p = whatever is decided on;
    pidAlgorithm->request->K_i = whatever is decided on;
    pidAlgorithm->request->K_d = whatever is decided on;
    pidAlgorithm->request->lastError = 0;
    pidAlgorithm->request->totalValue = 0;
    pidAlgorithm->request->T = 1000 / 100; // we are in a second / loop rate
}

// This point is made to handle the
void initSteps() {
    // EVERYTHING is just global here. This function is to be used in the actual code
    if (variance is not a number) {
        calculateVariance();
        set x, y, estimate, P, I, estimate = 0;
        call moveState;
        set lastPose = modelState->response->pose;
    } else { // variance is known -> find x, y, estimate
        calculate x, y, estimate using euclid;
        // Since we would have done the other branch in the first run
        // we do this before calling the moveState service
        set kalmanFilter->request->P, I, estimate =
            kalmanFilter->response->P, I;
    }
    call kalmanFilter;
    pidAlgorithm->request->error = kalmanFilter->response->P, I;
}

void defineMovement() {
    // EVERYTHING is just global here. This function is to be used in the actual code
    if (noisy) {
        moveSteps();
    } else {
        pidAlgorithm->request->error =
            sonarReadingSrv->response->readings->distance;
    }
}
```

```
sonarReadingSrv->response->readings->distance;
}
}
call PID_service;
movement.linear.x = pidAlgorithm->response.y;
pidAlgorithm->request->lastError =
    pidAlgorithm->request->error;
pidAlgorithm->request->totalValue =
    pidAlgorithm->response->totalValue;
pidAlgorithm->request->totalValue;
}

int main() {
    Init ros_node;
    ros::Rate rate = 100;

    define sonarReader (sonar wrapper service client);
    define driver (publisher for cmd_vel);
    if (noisy) {
        define kalmanFilter;
        declare lastPose;
        declare ModelStateSrv (storage for ModelState);
        define ModelStateSrv model_name = "turtlebot3_burger";
        define ModelState (service client);
    }
    declare pidAlgorithmSrv (storage for PID_service);
    define PID_service (node client for PID);
    definePIDInitialValues(pidAlgorithmSrv);
    bool firstPid = true;

    // This is the important section
    while (ros::ok()) {
        spin once;
        call sonarReadingSrv

        declare Twist movement;
        if (we are not facing the bowl) {
            define turn();
        } else {
            define movement();
        }
        // publish either the turn or the movement
        driver.publish(movement);
        rate.sleep();
    }
}
```

## 5.4 Implementation

Following the pseudocode, the controller was implemented as visible in figure 7. Note that the turn rate was set at ten degrees.

```
//include <iostream>
#include "PosVec.h"
#include "AssignmentSetup/Sonar.h"
#include "Assignment1/getSonarReadings.h"
#include "Assignment1/pid_algorithm.h"
#include "Assignment1/kalman_filter.h"
#include "Utils.h"
#include "Geometry_msgs/ twist.h"
#include "nav_msgs/Odometry.h"
#include "geometry_msgs/ Pose.h"
#include <vector>

// Maximum is 1.6 radians / second * half a radius
// One degree = 180/57.3 PT
const double TURNRATE = R_PI / 180 * 10;
//if(0) NOISY SONAR
uint32_t variancesamples = 1000;
double

double euclideanDistance(geometry_msgs::Pose p1, geometry_msgs::Pose p2) {
    ROS_INFO("Calculating euclidean distance: p1: %f %f %f p2: %f %f %f", p1.x, p1.y, p1.z, p2.x, p2.y, p2.z);
    p1.position.x, p1.position.y, p1.position.z,
    p2.position.x, p2.position.y, p2.position.z);
    return sqrt(
        pow(p1.position.x - p2.position.x, 2) +
        pow(p1.position.x - p2.position.y, 2) +
        pow(p1.position.x - p2.position.z, 2)
    );
}

};

void definturn(geometry_msgs::Twist &definturnOf,
               assignment1::getSonarReadings &sonarReadingServ) {
    // We need to move
    if (sonarReadingServ.response.readings.distance == UNINIT_MAX ||
        sonarReadingServ.response.readings.distance == UNINIT_MAX) {
        // If there is no data, we need to turn right
        definturnOf.angular.z = TURNRATE;
    } else if (sonarReadingServ.response.readings.distance == UNINIT_MAX) {
        // If there is some distance, we need to turn positive
        definturnOf.angular.z = TURNRATE;
    } else {
        // If there is enough distance, we need to turn negative
        definturnOf.angular.z = TURNRATE * -1;
    }
}

double calculateVariance(std::vector<uint16_t> &sonarReadings) {
    ROS_INFO("Number of values collected: %d", sonarReadings.size());
    double mean(0.0);
    for (auto &v : sonarReadings) {
        mean += v;
    }
    return mean / v;
```

```

54 mean = (sonarReadings.size());
55 ROS_INFO("x1f", mean);
56
57 double variance(0);
58 for (auto &v : sonarReadings) {
59     variance += (v - mean) * (v - mean);
60 }
61 variance /= sonarReadings.size() - 1;
62 ROS_INFO("variance x1f", variance);
63 return variance;
64 }
65
66 void defineIDrvInInitValues(assignment: pid_algorithm_t ApidAlgorithmIn) {
67     // set some parameters for the algorithm readability
68     pidAlgorithmIn.request.k_P = 0.25/100;
69     pidAlgorithmIn.request.k_I = 0;
70     pidAlgorithmIn.request.k_D = 0;
71     pidAlgorithmIn.request.integrator = 0;
72     pidAlgorithmIn.request.totalValue = 0;
73     pidAlgorithmIn.request.t = 1000 / 100; // ms in a second / loop rate
74 }
75
76 // Reading this would be incredibly painful because a lot of sections cannot
77 // be extracted thanks to how early it would be in main. As a workaround,
78 // check the main code in the report first before trying to understand this.
79 int main(int argc, char *argv[]) {
80     ros::init(argc, argv, "Controller");
81     ros::NodeHandle nodelandIn;
82     ros::Rate rate(100);
83
84     assignment1: getSonarReadings sonarReadingsIn;
85     ros::ServiceClient sonarClientIn(assignment: getSonarReadings({
86         "sonar_wrapper"
87     }));
88     ros::Publisher driver =
89         nodelandIn.advertise<geometry_msgs::Twist>("cmd_vel", 1000);
90
91     if(ros::IS_INIT) SONAR
92     assignment1: kalmInFilter kalmInFilterIn;
93     // consistency of variance to how new we have a check for whether it's been
94     // initialized later
95     kalmInFilterIn.request.R_I = std::numeric_limits<double>::quiet_NaN();
96     ros::ServiceClient kalmInFilter =
97         nodelandIn.serviceClient<geometry_msgs::Twist>("kalmInFilter");
98     "kalmInFilter";
99 }
100
101 geometry_msgs::Pose lastPose;
102 gazebo_msgs::GetModelState ModelStateIn;
103 ModelStateIn.request.model_name = "turtlebot3_burger";
104 ros::ServiceClient modelState =
105     nodelandIn.serviceClient<gazebo_msgs::GetModelState>{
106         "/gazebo/get_model_state"
107     }

```

```

90 };
91 #endif
92
93 assignment::pid_algorithm(pidAlgorithmSrv,
94 ros::ServiceClient PID_client) {
95     nodeHandle.serviceClient<assignment::pid_algorithm>(
96         "pid_algorithm")
97     );
98     definePIDsvrInitialValues(pidAlgorithmSrv);
99     bool firstPid = true;
100
101     while(ros::ok()) {
102         ros::spinOnce();
103         // Would love to extract all of these node client calls so they do not
104         // take as long time each, but the current function is to show...
105         if (!sonarReader.call(sonarReadingSrv)) {
106             ROS_ERROR("failed to use assignment/sonar_wrapper: is it running?");
107             rate.sleep();
108             continue;
109         }
110
111         geometry_msgs::Twist movement;
112         if (sonarReadingSrv.response.readings.distance == ULLIHI_MAX) {
113             // We need to turn + extract it as a function for cleanliness
114             defineTurn(movement, sonarReadingSrv);
115         } else {
116             // We can start driving forwards
117             if(!firstPid){
118                 // don't go ahead and generate the variance now
119                 // don't extract method because we use rate.
120                 if((kalmefilterSrv.request.R_1)) {
121                     std::vector<int> sonarSamples;
122                     while(ros::ok() && variancesamples > 0){
123                         variancesamples--;
124                         if(variancesamples % 5 != 0){
125                             ROS_INFO("collecting samples for variance calculation...".c_str(), variancesamples);
126                         }
127                         ros::spinOnce();
128                         if(sonarReader.call(sonarReadingSrv)){
129                             ROS_ERROR("failed to use assignment/sonar_wrapper: is it running?");
130                             rate.sleep();
131                             continue;
132                         }
133                         sonarSamples.push_back(sonarReadingSrv.response.readings.distance);
134                         rate.sleep();
135                     }
136                     kalmefilterSrv.request.R_1 = calculateVariance(sonarSamples);
137                     // since this is the first run x,y,z,p_x_estimate and
138                     // p_y_estimate = 0 As at zero offset this estimate and
139                     // P_P = 0.0 As P_z = (1/3)*P_x_estimates and K = 0 when
140                     // p_x_estimates = 0.
141                     kalmefilterSrv.request.z_f = 0;
142                     kalmefilterSrv.request.y_p_estimate = 0;
143                     kalmefilterSrv.request.p_x_estimate = 0;
144                     kalmefilterSrv.request.p_z_estimate = 0;
145
146                     if(modeState.call(modeStateSrv)){

```

```

514         // (ModelState.call(ModelState.v)) {
515             ROS_ERROR("Failed to use gazebo/get_model_state: is it running?");
516             rate.sleep();
517             continue;
518         }
519         // variance has been calculated, do the actual filter instead
520         // (ModelState.call(ModelState.v)) {
521             ROS_ERROR("Failed to use gazebo/get_model_state: is it running?");
522             rate.sleep();
523             continue;
524         }
525         lastPose = ModelStateSrv.response.pose;
526     } else {
527         // variance has been calculated, do the actual filter instead
528         // (ModelState.call(ModelState.v)) {
529             ROS_ERROR("Failed to use gazebo/get_model_state: is it running?");
530             rate.sleep();
531             continue;
532         }
533         lastPose = ModelStateSrv.response.pose;
534         kalmanFilterSrv.request.y_i.estimate =
535             euclideanDistance(lastPose, ModelStateSrv.response.pose);
536         // Also want P estimate for the next run
537         kalmanFilterSrv.request.P_i.estimate =
538             kalmanFilterSrv.response.P_i;
539     }
540     kalmanFilterSrv.response.P_i;
541 }
542
543 kalmanFilterSrv.request.P_i = sonarReadingSrv.response.readings.distance;
544 // (kalmanFilter.call(kalmanFilter.v)) {
545     ROS_ERROR("Failed to call kalman filter");
546     rate.sleep();
547     continue;
548 }
549
550 pidAlgoithrSrv.request.error = kalmanFilterSrv.response.y_i;
551
552 #else
553 pidAlgoithrSrv.request.error = sonarReadingSrv.response.readings.distance;
554
555 #endif
556
557 ROS_INFO("read error as %f", (longint) (int) pidAlgoithrSrv.request.error);
558 // (PID.service.call(pidAlgoithrSrv)) {
559     ROS_ERROR("Failed to use assignment/sonar_wrapper: is it running?");
560     rate.sleep();
561     continue;
562 }
563
564 movement.linear.x = pidAlgoithrSrv.response.y;
565 // Save data request based on the responses for the next run
566 pidAlgoithrSrv.request.lastError =
567     pidAlgoithrSrv.request.error;
568 pidAlgoithrSrv.request.totalValue =
569     pidAlgoithrSrv.response.totalValue;
570 }
571
572 driver.publish(movement);
573 ROS_INFO("Published linear x: %f angular z: %f",
574     movement.linear.x, movement.angular.z);
575
576 rate.sleep();
577
578 return EXIT_SUCCESS;

```

Figure 7: Implementation of the controller. Read as top left -; top right -; bottom left -; bottom right



## References

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